



Nice Ride: Data-Driven Dynamics



Minneapolis Bike Share Company

Presented by
The Fundamental Group



Why We're Here

- Nice Ride has become an integral part of the the Twin Cities



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- No one likes an empty or a full station



Why We're Here

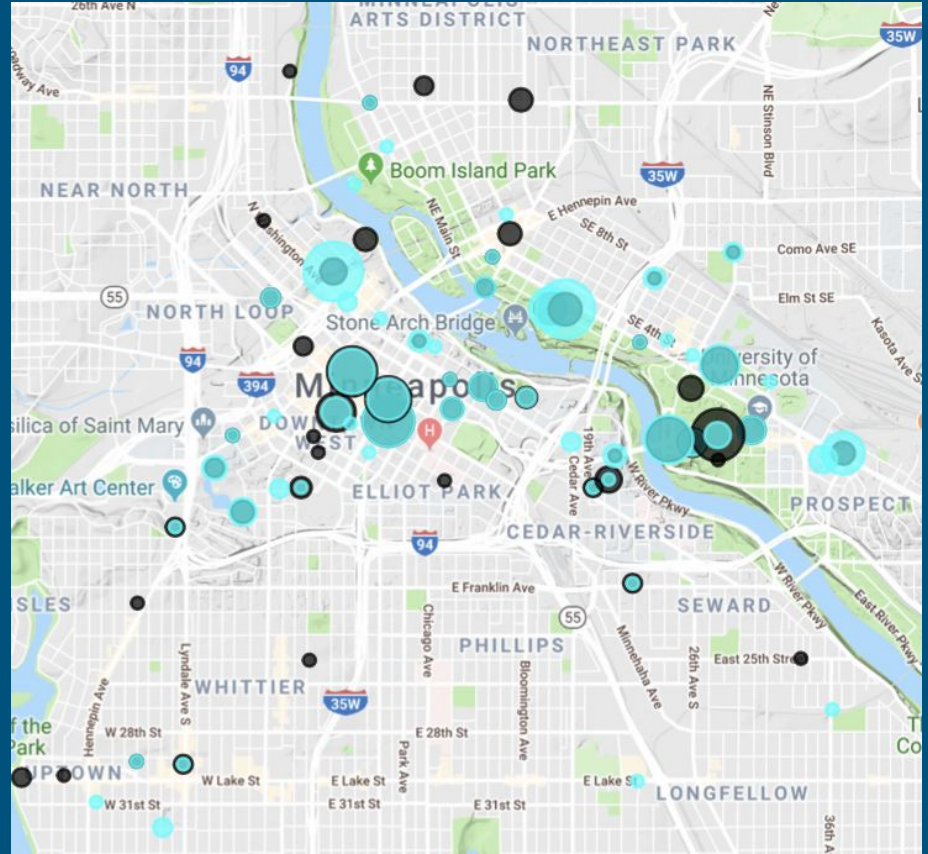
- Nice Ride has become an integral part of the the Twin Cities
- No one likes an empty or a full station
- We introduce automated, predictive, data-driven solutions to solve this problem



The Problem

Certain stations lose or gain bikes quickly

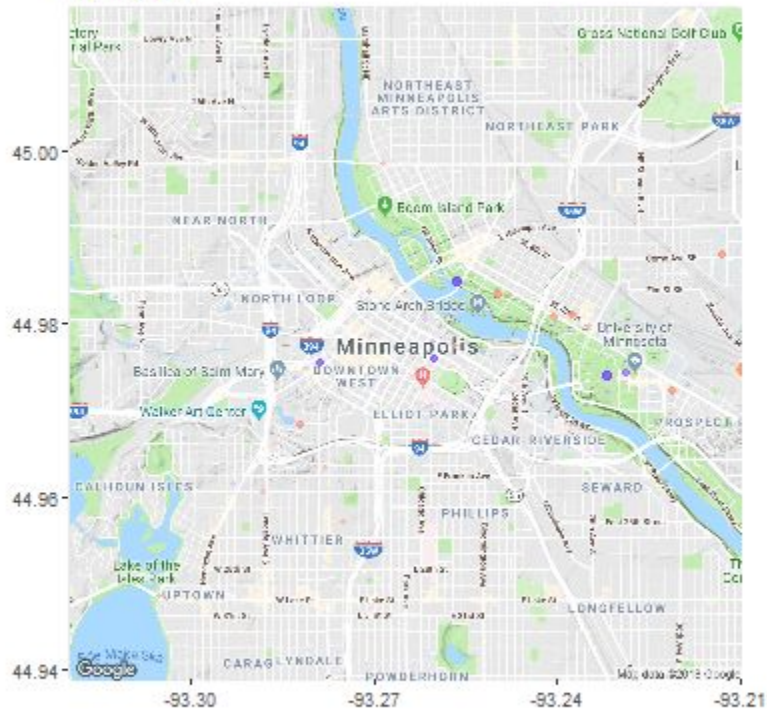
- This leaves stations empty or full for long durations of time
- This can cause Nice Ride to **lose customers**



Blue - Bikes lost
Black - Bikes Gained

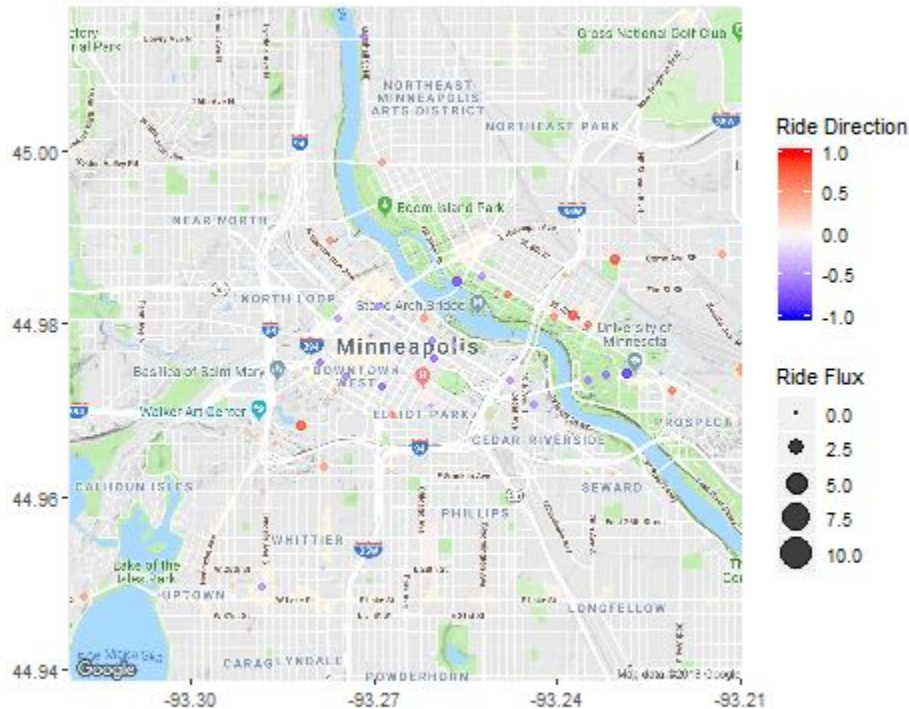
Average Ride Flow Patterns

Mon, 0:00



Monday

Sat, 0:00



Saturday

Our Goals

1. Predict the hourly change in bikes at each Nice Ride station

2. Predict overall system demand on any given day



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graph TD; A[1. Predict the hourly change in bikes at each Nice Ride station] --> C[3. Use this to develop a strategy for preventing the imbalance of bicycles to maximize customer satisfaction]; B[2. Predict overall system demand on any given day] --> C;
```

3. Use this to develop a strategy for preventing the imbalance of bicycles to **maximize customer satisfaction**

Predicting Demand

1. **Hourly change in bikes per station**
(accurate within ± 1.4 bikes per hour)

2. **Daily System Demand:**
Linear Regression Model

- Temperature
- Precipitation
- Average Wind Speed
- Season
- Thunder/Fog



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```
graph TD; A[1. Hourly change in bikes per station] --> C[3. Predict and prevent bike imbalance]; B[2. Daily System Demand: Linear Regression Model] --> C;
```

3. **Predict and prevent bike imbalance**

Current Situation

- Bike shuttling occurs, but is not automated
- Currently a couple of guys with trailers making up routes on the fly
- Reactive rather than proactive
- Many stations are often empty or full!



Dynamic Solution to the Commuter Problem

Set-up:

- Drivers start at Nice Ride Headquarters and use a truck to deliver bikes

Algorithm:

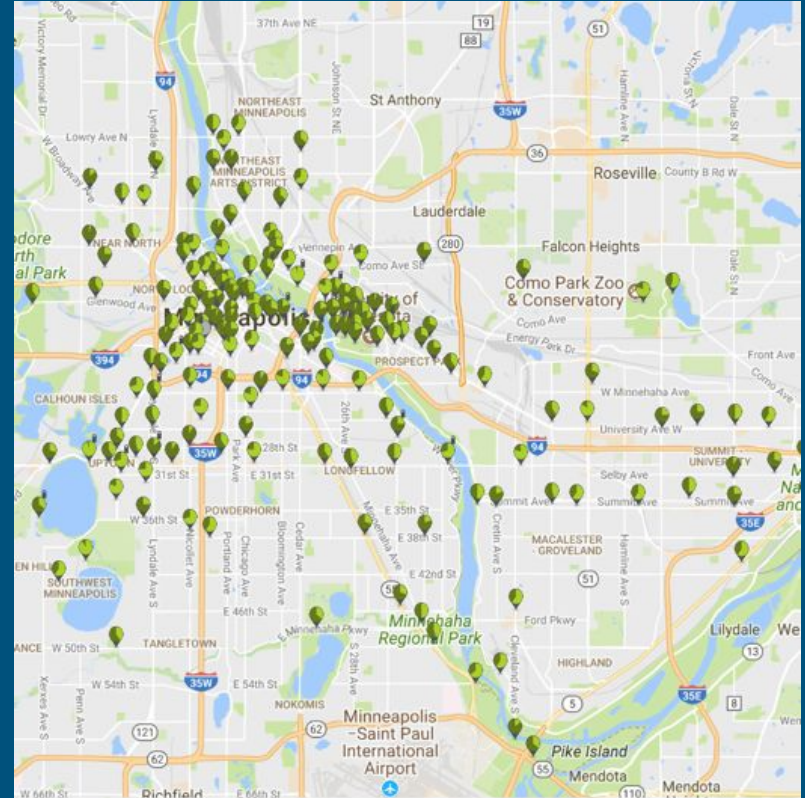
- **Input:** real-time number of bikes and desired number of workers
- **Output:** morning delivery schedule, including bikes needed
 - Rinse and repeat for afternoon and evening

All you have to do is press a button on your laptop!

How Our Algorithm Works

1. Pick a time period during which to deliver bikes (e.g. from 6am to 10am on Monday morning)

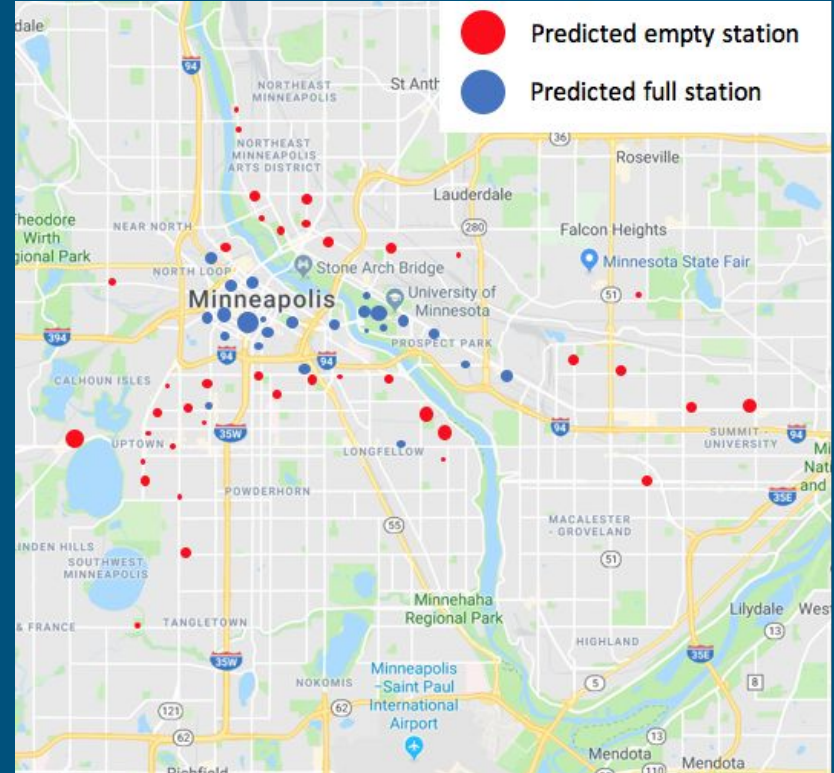
- Pull real-time number of bikes at each station at 6am.
 - *Nice Ride already has this data available online*



How Our Algorithm Works

2. **Predict** which stations will have too many or too few bikes at 10 am.

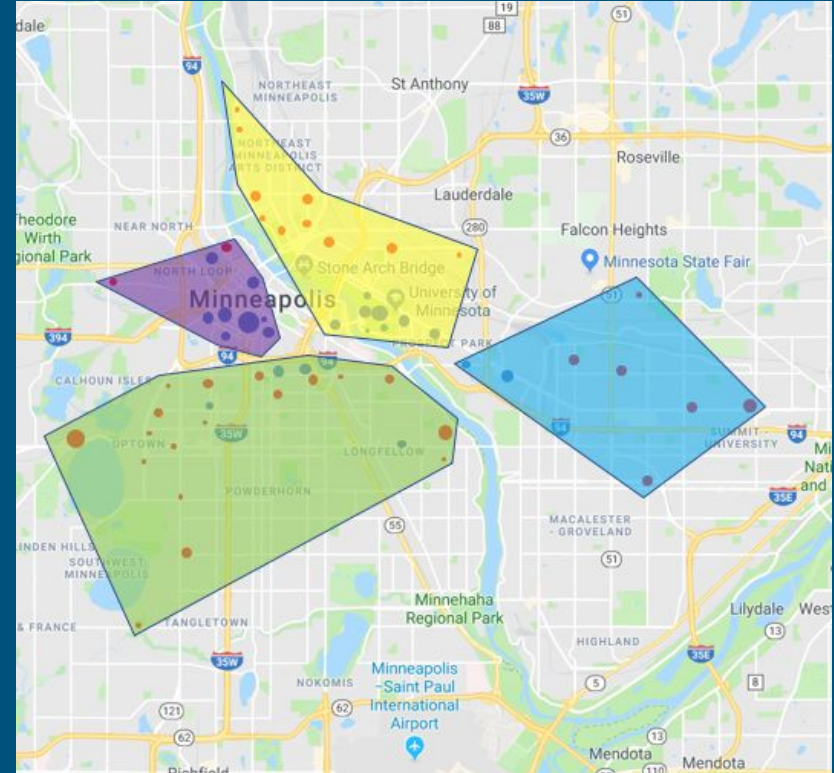
- If this number is lower than 20% or greater than 80% of capacity, the algorithm flags the station and calculates the number of bikes to add or remove.



How Our Algorithm Works

3. **Group** the flagged stations into a delivery zone for each worker.

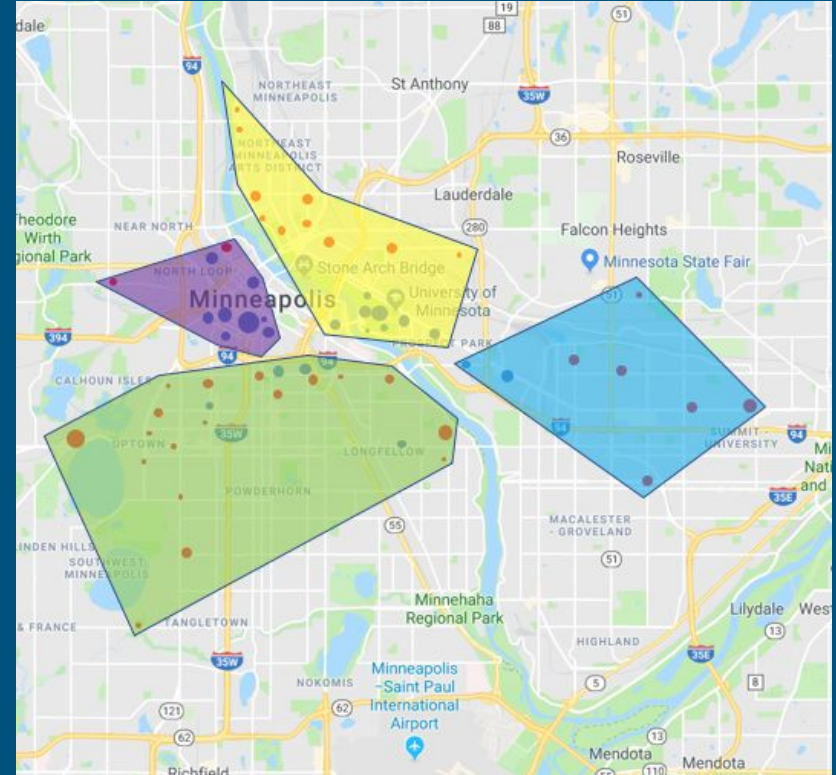
- The algorithm automatically creates zones using k-means clustering
 - Minimizes the distance between stations in a zone.



How Our Algorithm Works

4. Minimize distance traveled in each workers' zone and produce optimal delivery schedule

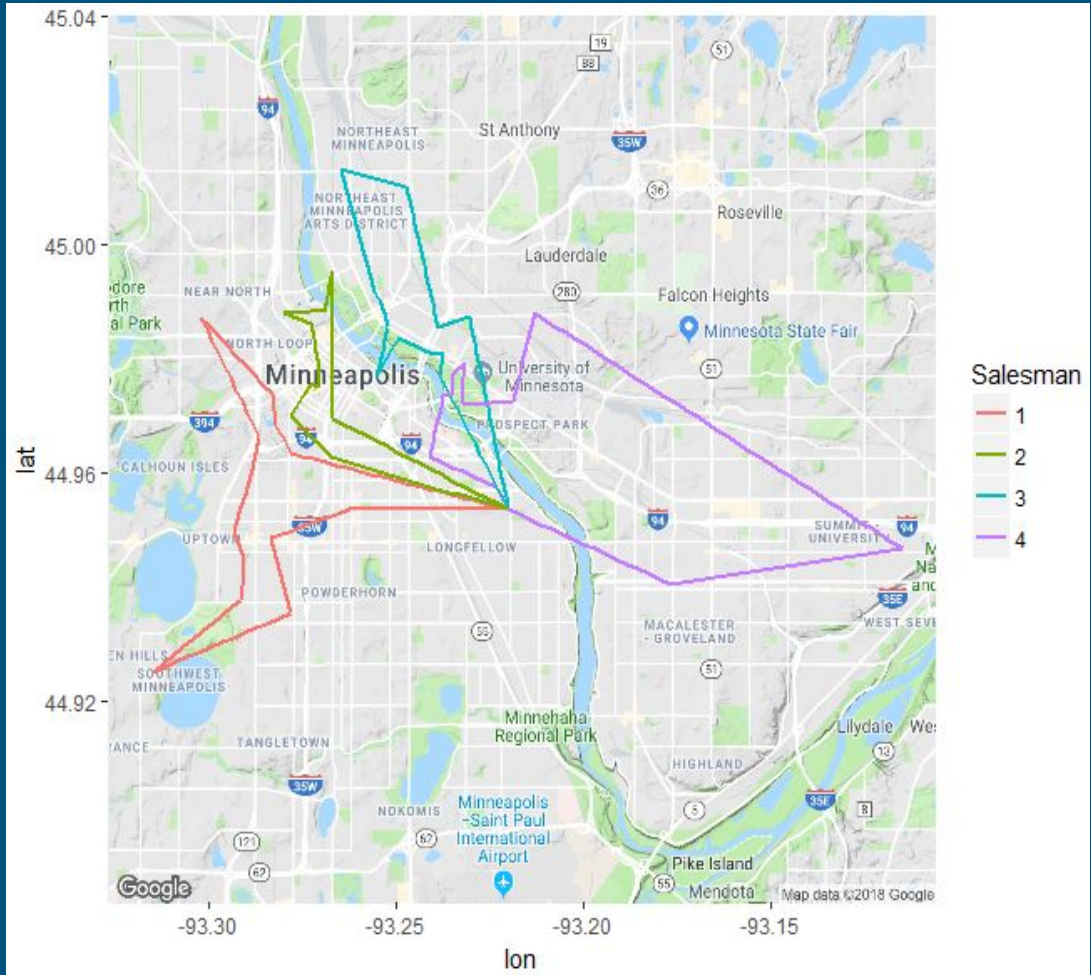
- The algorithm runs the traveling salesman problem
- Produces an optimal delivery schedule in each zone that minimizes distance traveled
 - *Time and cost minimized*



Monday Morning

Example: four vans originating at Nice Ride Headquarters

- Our algorithm produces **fully automated** morning delivery schedules based on **real-time data**
- **Minimizes** time spent by workers
- **Maximizes** customer satisfaction

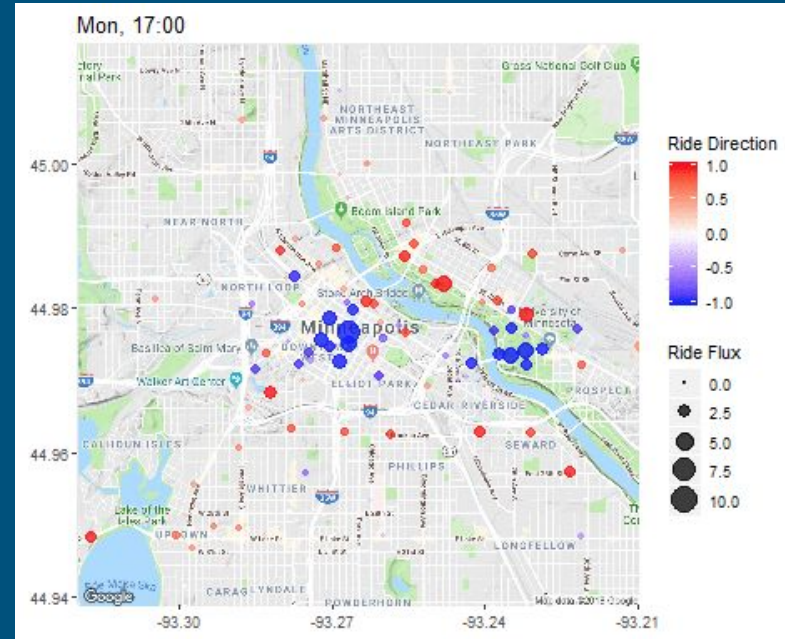


What we bring to the Nice Ride Family

- Our solution represents a significant improvement to the current Nice Ride system
- Can we **save money** by also implementing long-term solutions that minimize the number of workers needed to deliver bikes?

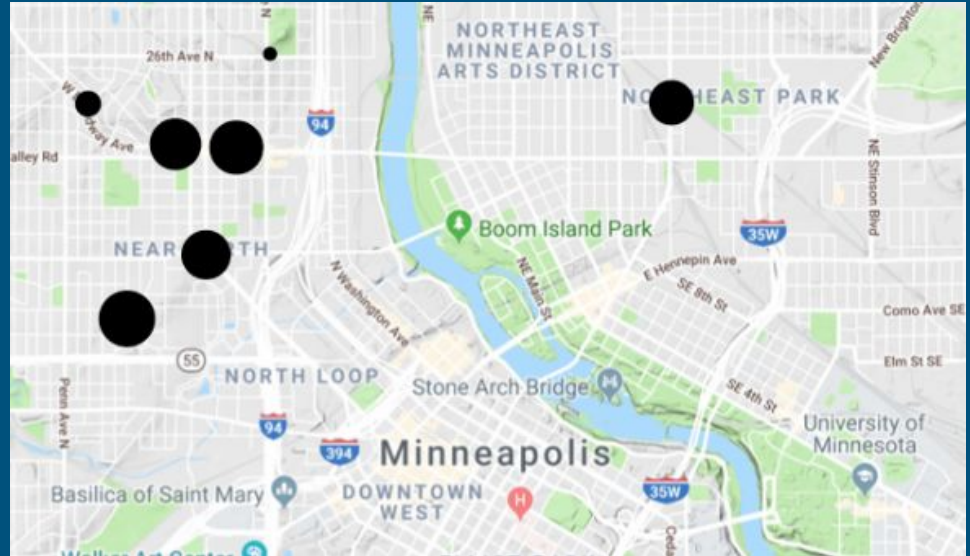
Further Solutions and Areas to Investigate

- Build new stations near high demand areas
 - Investigate expansion around the U of M



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 - Investigate expansion around the U of M
- Shift docks from low-demand stations (pictured) to high-demand stations



Our Final Recommendations

Action: Implement algorithm to automate bike shuttling

Consequences: Improved customer satisfaction

No stations left empty or full

Fewer worker hours

Future Work: Investigate where to build new stations

Investigate how to reshuffle docks

Thank You!

The Fundamental Group:

Ariel Bowman

Olivia Cannon

Michael Ramsey

Eric Roberts

Elizabeth Wicks

Sheng Zhang

Questions?

Thank You! Questions?

The Fundamental Group:

Michael Ramsey: Duke of Data

Eric Roberts: King of Kitty Kat the Almighty

Olivia Cannon: Princess of Presentation, Duchess of Dynamics

Ariel Bowman: Lady of Linear Models, Empress of Elevator Pitches

Elizabeth Wicks: Queen of Quantitative Modeling, Nth of Her Name

Sheng Zhang: Prince of Python